Supplemental Text S3.

In silico evolution

Our simulations employ an ODE model of the switch (Figure 6A) based on individual reactions between the components, Kinase1, Kinase2, and Inhibitor, which are analogous to Cln1/2-Cdk1, Clb5/6-Cdk1, and Sic1 in the yeast G1/S switch. We modeled Inhibitor with six phosphosites, all of which required to be phosphorylated for its rapid degradation. Multisite phosphorylation of Inhibitor was assumed to occur in an ordered fashion, from site #1 through site #6. The reaction set is given below.

```
# KINASE1 and KINASE2 synthesis
kinase1 → kinase1 + KINASE1, Hill(t, k_s1, K_s1, n_s1)
kinase2 → kinase2 + KINASE2, Hill(t, k_s2, K_s2, n_s2)

# KINASE1 AND KINASE2 degradation
KINASE1 → Ø, k_d1
KINASE2 → Ø, k_d2

# INHIBITOR synthesis
inhibitor → inhibitor + INHIBITORp_0, k_s3

# Phosphorylation of free INHIBITOR by KINASE1
INHIBITORp_i + KINASE1 → KINASE1 + INHIBITORp_{i+1}, k_{1,i}

# Phosphorylation of INHIBITOR-KINASE2 complex by KINASE1
INH_K2p_i + KINASE1 → KINASE1 + INH_K2p_{i+1}, k_{1,i}

# Phosphorylation of free INHIBITOR by KINASE2
INHIBITORp_i + KINASE2 → KINASE2 + INHIBITORp_{i+1}, k_{2,i}

# Phosphorylation of INHIBITOR-KINASE2 complex by KINASE2
INH_K2p_i + KINASE2 → KINASE2 + INH_K2p_{i+1}, k_{2,i}

# Dephosphorylation of free INHIBITOR
INHIBITORp_{i+1} → INHIBITORp_i, k_3

# Dephosphorylation of INHIBITOR-KINASE2 complex
INH_K2p_{i+1} → INH_K2p_i, k_3

# Sequestration of free KINASE2 by free INHIBITOR
INHIBITORp_j + KINASE2 → INH_K2p_j, k_4

# Dissociation of INHIBITOR-KINASE2
INH_K2p_j → INHIBITORp_j + KINASE2, k_5

# Degradation of free INHIBITOR
```
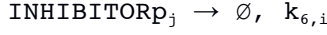
```
INHIBITORp_j  →  ∅,  k_{6,i}
```

Degradation of INHIBITOR-KINASE2 complex to free KINASE2
```
INH_K2p_j  →  KINASE2,  k_{6,i}
```

Degradation of INH_K2 to free INHIBITOR
```
INH_K2p_j  →  INHIBITORp_j,  k_{d2}
```

Here, lowercase symbols, `kinase1`, `kinase2`, and `inhibitor` denote genes; whereas uppercase symbols, `KINASE1`, `KINASE2`, and `INHIBITORp_i` denote proteins. `INH_K2p_i` represents the Inhibitor-Kinase2 complex. Rate of each reaction is given after the comma. The reactions for `KINASE1` and `KINASE2` synthesis employ Hill functions to produce sigmoidal curves. These reactions are implemented as a function of time. The reaction rate is calculated using the function $\text{Hill}(t, k, K, n) = k \frac{t^n}{K^n + t^n}$, where t represents time in seconds. The indices, i and j are integers denoting the number of phosphorylated sites ($0 \leq i \leq 5$ and $0 \leq j \leq 6$). In the 1P model, i=0 and j=0, 1. We used molecule counts in the simulations (not concentrations), and chose k_{s1}=k_{s2}=160, k_{d1}=k_{d2}=0.08, K_{s1}=K_{s2}=10, n_{s1}=n_{s2}=4, k_{s3}=23.1, k_3=0.5, k_4=0.5, k_5=0.24, $k_{6,i<6}$=0.0154, $k_{6,6}$=0.17. Constants, k_{s1}, k_{s2}, k_{d1}, k_{d2}, K_{s1}, K_{s2}, n_{s1}, and n_{s2} are subject to extrinsic noise (to simulate cell to cell variability). Each time a copy of the circuit is run, the specified reaction rate is multiplied by 1+ρ, where ρ is a random number uniformly distributed on (-x, x). x is chosen as 0.2 for k_{s1}, k_{s2}, K_{s1}, and K_{s2}; and 0.7 for n_{s1} and n_{s2}. The phosphorylation rates, $k_{1,i}$, and $k_{2,i}$ are real numbers between [10^{-10}, 0.1] and are subject to mutations. At the beginning of the simulation, they are assigned random numbers from a log-uniform distribution on [10^{-6}, 0.1]. Dephosphorylation was assumed to occur at a constant rate. Mutations are modeled by multiplying the reaction rates by a random number uniformly distributed on (0, 2]. Mutation rate per reaction rate per generation was chosen as 0.3. Mutations are carried from one generation to the next, unlike the extrinsic noise, which is not heritable. $k_{1,i}$= 0 in the double-negative feedback loop simulations, and $k_{2,i}$= 0 in the linear circuit.

We define timing of the Kinase2 activation as the time the number of Kinase2 exceeds the number of Inhibitor. We take sharpness as the slope of the free Kinase2 curve at its half-maximum level.

Optimization by *in silico* Evolution

The algorithm resembles evolution by natural selection, and works as follows.

1. **Initialization**. $N=1000$ copies of the network is generated. Phosphorylation rates of each copy (realization) are assigned random rates as discussed above. Initial protein counts are all zero except Inhibitor=1500. All genes have exactly one copy each.

2. **Scoring**. Reaction rates that are subject to extrinsic noise are modified as described above. (In this set, only Kinase1 and Kinase2 production/degradation are subject to noise.) Then, the network is run, and a fitness score is calculated based on the output. Fitness score for timing is

$$f_s = (t_{\text{activation}} - t_{\text{desired}})^2.$$

Here t_{desired} denotes desired activation time for Kinase2. Sharpness score is simply the sharpness of the free Kinase2 curve as defined above. Lower scores mean higher fitness.

3. **Elimination and duplication.** Half of the population with lower fitness is eliminated. The rest is duplicated to make up the deficit.

4. **Mutation.** Duplicates are mutated as described above. Mutations only affect phosphorylation rates (*i.e.* catalytic efficiencies).

Steps 2-4 are repeated for $m=1000$ generations. Longer runs were not observed to change the final distributions of the phosphorylation rates qualitatively.

Simulations were written in C++ for distributed-memory parallel processing using OpenMPI (Gabriel et al., 2004). Sundials libraries were used for integration (Hindmarsh et al., 2005). The source code is available on request.

Supplemental References

Gabriel, E., Fagg, G.E., Bosilca, G., Angskun, T., Dongarra, J.J., Squyres, J.M., Sahay, V., Kambadur, P., Barrett, B., Lumsdaine, A., et al. (2004). Open MPI: Goals, concept, and design of a next generation MPI implementation. In In Proceedings, 11th European PVM/MPI Users' Group Meeting, pp. 97–104.

Hindmarsh, A.C., Brown, P.N., Grant, K.E., Lee, S.L., Serban, R., Shumaker, D.E., and Woodward, C.S. (2005). SUNDIALS: Suite of nonlinear and differential/algebraic equation solvers. ACM Trans. Math. Softw. *31*, 363–396.